

Differential Equations

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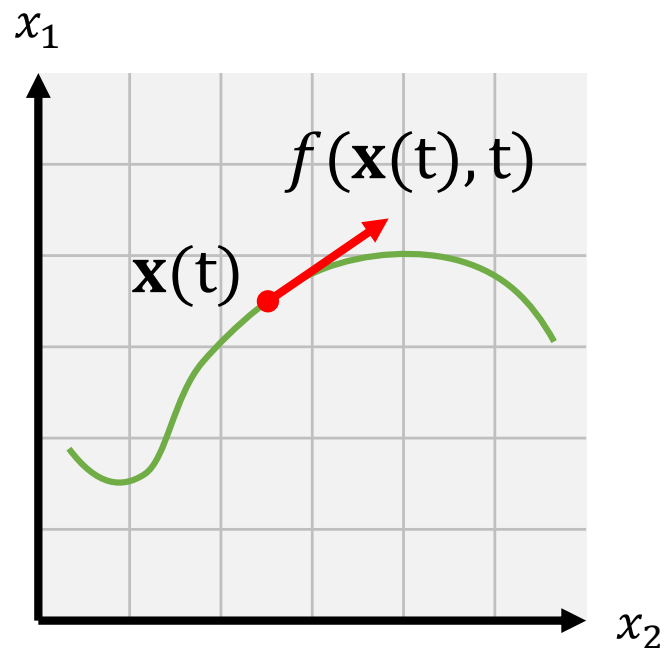
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Physically Based Modeling (SIGGRAPH 2001 course) by Andrew Witkin

Differential Equations

- It describes the relation between an unknown function and its (first, second, partial, etc) derivatives
- Many natural phenomena can be explained by using differential equations
 - Newton's equation (motions of macroscopic objects)
 - Diffusion equation (heat transfer)
 - Maxwell equation (electromagnetism)
 - Navier-Stokes equation (motions of fluids)

A Canonical Differential Equation

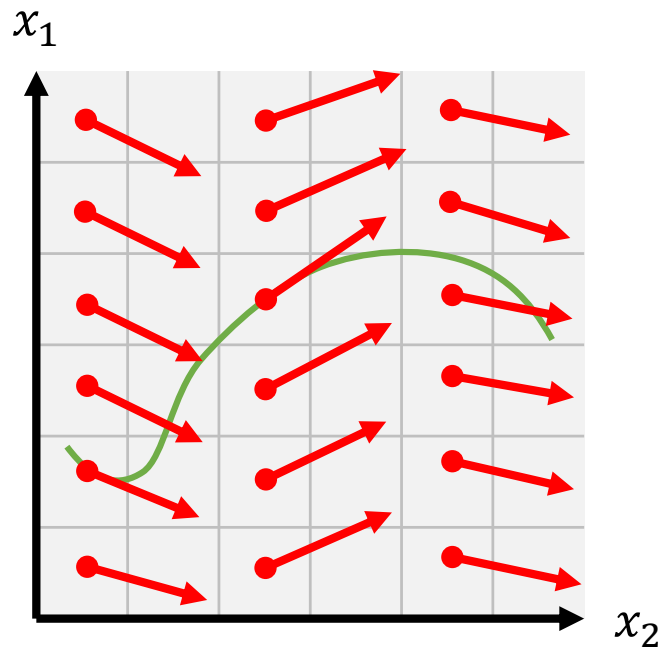


$$\dot{\mathbf{x}}(t) = f(\mathbf{x}(t), t)$$

$\mathbf{x}(t)$: a state of the system

$f(\mathbf{x}(t), t)$: the time derivative of $\mathbf{x}(t)$

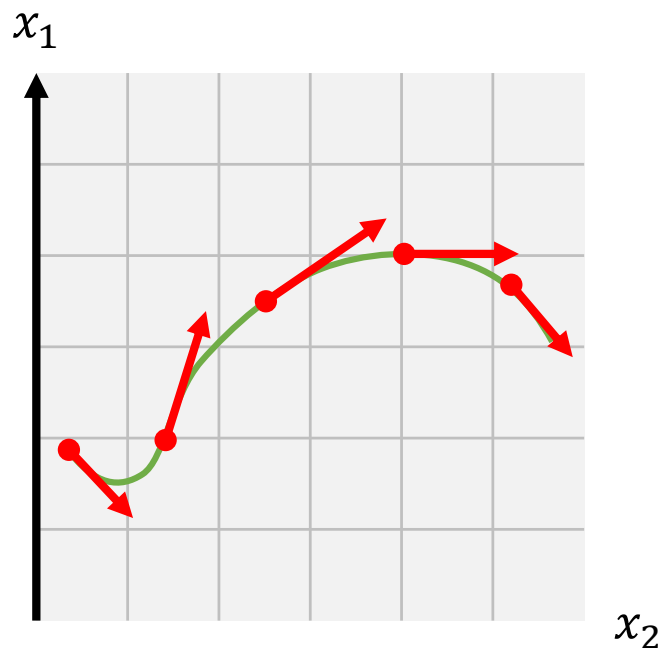
Vector Field



$$\dot{\mathbf{x}}(t) = f(\mathbf{x}(t), t)$$

- The differential equation defines a vector field regarding the velocity
- If $\mathbf{x}(t)$ is a location of moving point \mathbf{p} at time t , the equation tells us that the point should have the same velocity to $f(\mathbf{x}, t)$ given its location $\mathbf{x}(t)$

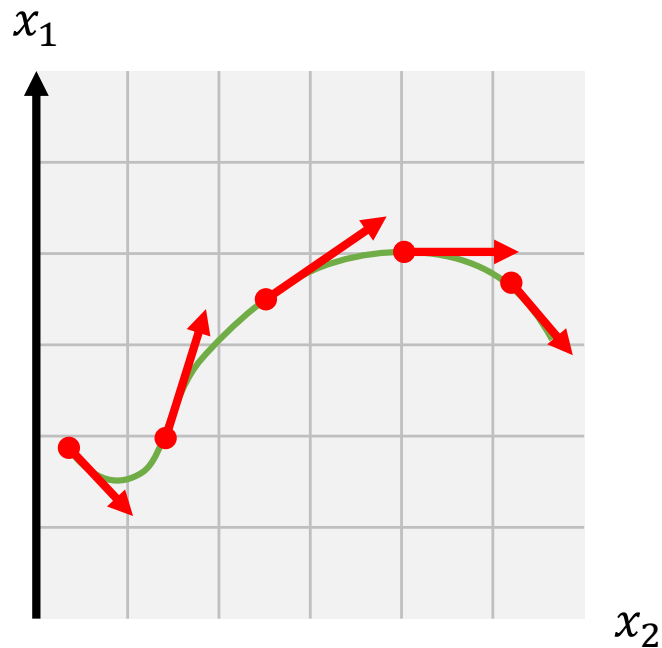
Integral Curves



$$\dot{\mathbf{x}}(t) = f(\mathbf{x}(t), t)$$

- We can think f as driving \mathbf{p} from point to point, like an ocean current
- Once a point \mathbf{p} is dropped at the location \mathbf{x} at time t , all future motion is fully determined by f
- The trajectory swept out by \mathbf{p} through f forms an **integral curve** of the vector field

Initial Value Problems

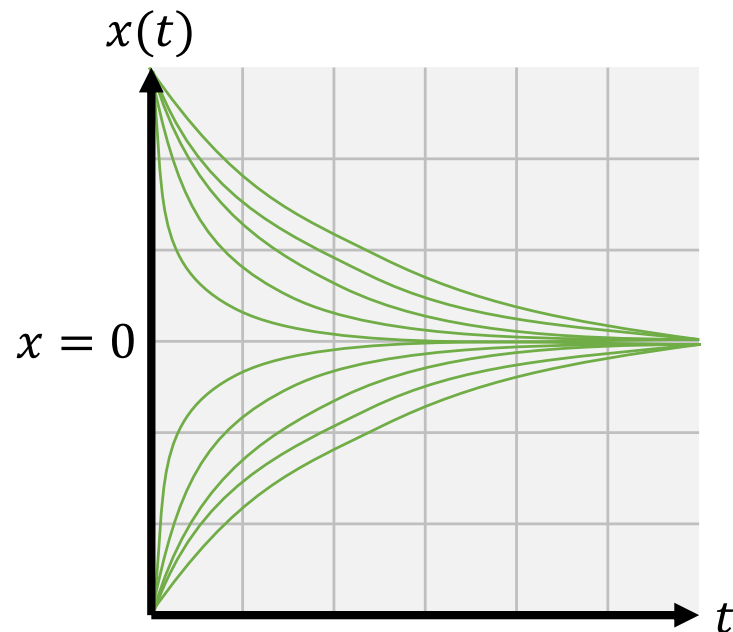


$$\dot{\mathbf{x}}(t) = f(\mathbf{x}(t), t)$$

- Given the starting point, find the integral curve
- Note that ***when to start*** in addition to ***where to start*** could also affect the results if f depends directly on time t

Analytical Methods

- Ways to get analytical solutions
 - Substitution
 - Series expansion
 - Laplace transform



$$\dot{x}(t) = -kx(t)$$

$$\therefore x(t) = e^{-kx(t)}$$

Numerical Methods

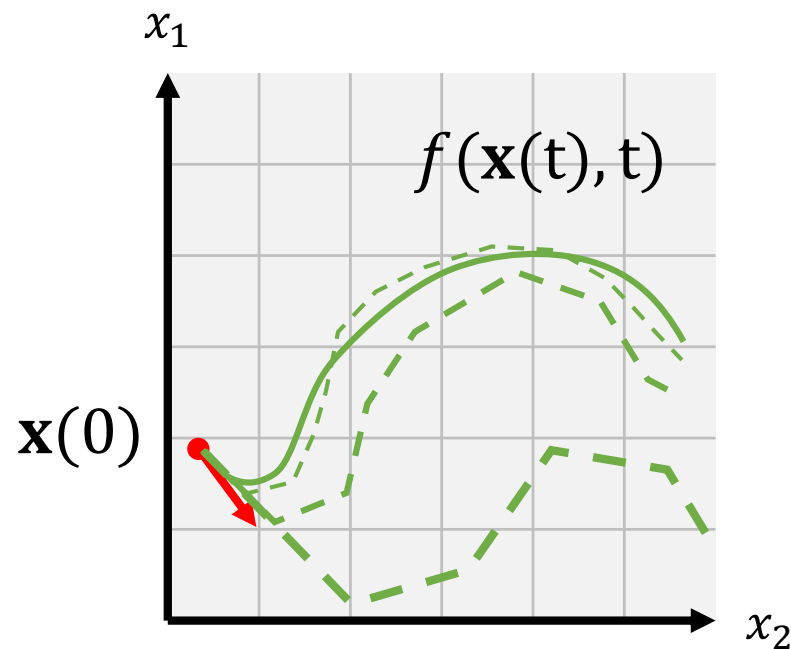
- As f gets complex, it gets harder to find analytical solutions
- Instead, we can use numerical methods to compute approximate integral curves
- The idea is to compute the change of state $\Delta \mathbf{x}$ over the fixed time interval $h = \Delta t$ by using $\dot{\mathbf{x}} = f$ (i.e. instantaneous velocity)

Euler's Method

$$\mathbf{x}(t + h) = \mathbf{x}(t) + \underline{h\dot{\mathbf{x}}}$$

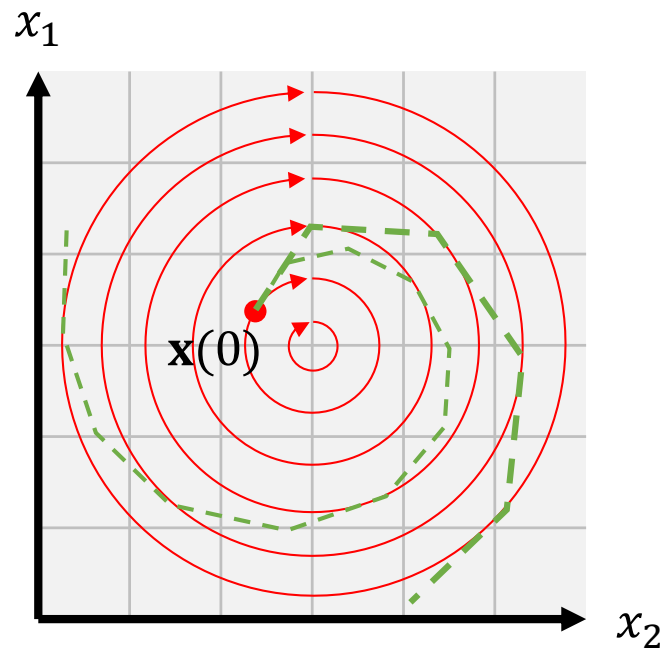
It assumes that the velocity does not change during the timestep h

$$\mathbf{x}(t + h) = \mathbf{x}(t) + hf(\mathbf{x}(t), t)$$



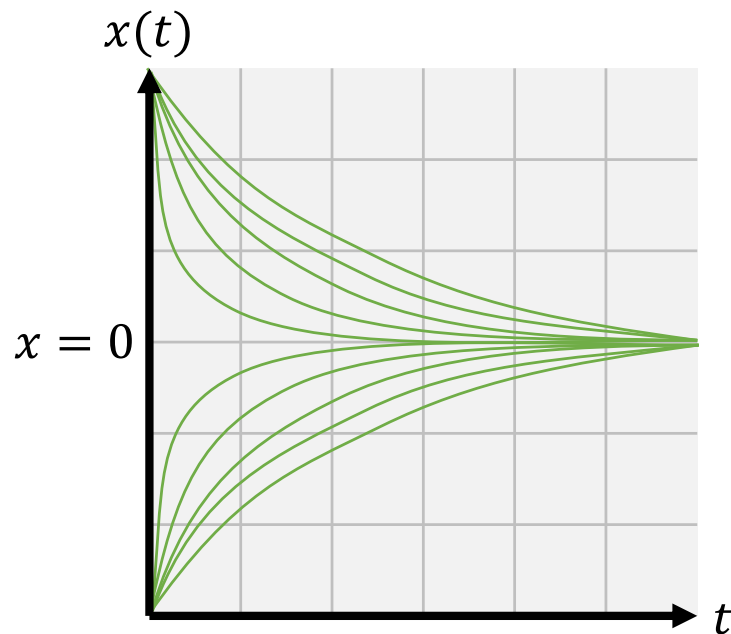
- Simplest numerical method
- Discrete time steps
- Bigger steps, bigger errors
- Smaller steps, larger computation

Problem 1: Inaccuracy



- Suppose we have a vector field rotating around with respect to a point, where any (exact) integral curves should be concentric circles
- When using Euler's method, the radius of integral curve gets larger and larger (i.e. spiral curves) regardless of timestep size

Problem 2: Instability



$$\dot{x}(t) = -kx(t)$$

$$x(t+h) = x(t) + h(-kx(t))$$

- When $h > 1/k$, we have $|\Delta x| = |x|$, so the solution oscillates around zero
- When $h \geq 2/k$, the solution does not converge
- Beyond $h > 2/k$, the oscillation diverges, and the system blows up

Euler's Method: Taylor Series

$$\mathbf{x}(t_0 + h) = \underbrace{\mathbf{x}(t_0) + h\dot{\mathbf{x}}(t_0)}_{\text{Euler's method}} + \underbrace{\frac{h^2}{2!}\ddot{\mathbf{x}}(t_0) + \dots + \frac{h^n}{n!}\frac{\partial^n \mathbf{x}(t_0)}{\partial t^n}}_{\text{Error occurred due to linear approximation } O(h^2)}$$

- Suppose we take steps of size h/m , this will produce about $(h/m)^2$ error at each timestep, and we need to take m steps more to simulate the original duration
- In total, the error is reduced **linearly** as we decrease the timestep

$$(h)^2 \cdot 1 = h^2 \quad \longrightarrow \quad \left(\frac{h}{m}\right)^2 \cdot m = \frac{h^2}{m}$$

The Mid-point Method

$$\mathbf{x}(t_0 + h) = \mathbf{x}(t_0) + h\dot{\mathbf{x}}(t_0) + \frac{h^2}{2!}\ddot{\mathbf{x}}(t_0) + \cdots + \frac{h^n}{n!} \frac{\partial^n \mathbf{x}(t_0)}{\partial t^n} + O(h^3)$$

$$\ddot{\mathbf{x}} = \frac{\partial f}{\partial \mathbf{x}} \frac{\partial \mathbf{x}}{\partial t} + \frac{\partial f}{\partial t}$$

Complicated to evaluate!

$$f(\mathbf{x}_0 + \Delta \mathbf{x}, t_0 + \Delta t) = f(\mathbf{x}_0, t_0) + \Delta \mathbf{x} \frac{\partial f}{\partial \mathbf{x}} + \Delta t \frac{\partial f}{\partial t} + \cdots$$

$$f\left(\mathbf{x}_0 + \frac{h}{2} \dot{\mathbf{x}}, t_0 + \frac{h}{2}\right) = f(\mathbf{x}_0, t_0) + \frac{h}{2} \left(\frac{\partial \mathbf{x}}{\partial t} \frac{\partial f}{\partial \mathbf{x}} + \frac{\partial f}{\partial t} \right) + O(h^2)$$

$$\frac{h^2}{2} \ddot{\mathbf{x}}(t) + O(h^3) = h \left(f\left(\mathbf{x}_0 + \frac{h}{2} \dot{\mathbf{x}}, t_0 + \frac{h}{2}\right) - f(\mathbf{x}_0, t_0) \right)$$

$$\Delta \mathbf{x} = \frac{h}{2} \dot{\mathbf{x}}(t_0) = \frac{h}{2} \frac{\partial \mathbf{x}}{\partial t}$$

$$\Delta t = \frac{h}{2}$$

Rearrange &
Multiply h both sides

The Mid-point Method

$$\boxed{\frac{h^2}{2} \ddot{\mathbf{x}}(t) + O(h^3)} = h \left(f \left(\mathbf{x}_0 + \frac{h}{2} f, t_0 + \frac{h}{2} \right) - f(\mathbf{x}_0, t_0) \right)$$

$$\mathbf{x}(t_0 + h) = \mathbf{x}(t_0) + h\dot{\mathbf{x}}(t_0) + \boxed{\frac{h^2}{2!} \ddot{\mathbf{x}}(t_0) + \dots + \frac{h^n}{n!} \frac{\partial^n \mathbf{x}(t_0)}{\partial t^n}}$$

$$\mathbf{x}(t_0 + h) = \mathbf{x}(t_0) + h\cancel{\dot{\mathbf{x}}(t_0)} + h \left(f \left(\mathbf{x}_0 + \frac{h}{2} f, t_0 + \frac{h}{2} \right) - f(\cancel{\mathbf{x}_0}, t_0) \right) + O(h^3)$$

$$\mathbf{x}(t_0 + h) = \mathbf{x}(t_0) + hf \left(\mathbf{x}_0 + \frac{h}{2} f(\mathbf{x}_0, t_0), t_0 + \frac{h}{2} \right) + O(h^3)$$

Mid-point method

The Mid-point Method

- The mid-point method is a 2nd-order method, (i) an euler step is computed, then (ii) the derivative is evaluated again at the step's midpoint which is used to calculate the step

$$\mathbf{x}(t_0 + h) = \mathbf{x}(t_0) + hf\left(\mathbf{x}_0 + \frac{h}{2}f(\mathbf{x}_0, t_0), t_0 + \frac{h}{2}\right)$$

- The error is reduced ***quadratically*** as we decrease the timestep

$$(h)^3 \cdot 1 = h^3 \quad \longrightarrow \quad \left(\frac{h}{m}\right)^3 \cdot m = \frac{h^3}{m^2}$$

Higher-order Methods

- By evaluating f a few more times, we can eliminate higher and higher orders of derivatives. The most popular procedure for doing this is a method called **Runge-Kutta** of order 4 and has an error per step of $O(h^5)$

$$k_1 = hf(\mathbf{x}_0, t_0)$$

$$k_2 = hf\left(\mathbf{x}_0 + \frac{k_1}{2}, t_0 + \frac{h}{2}\right)$$

$$k_3 = hf\left(\mathbf{x}_0 + \frac{k_2}{2}, t_0 + \frac{h}{2}\right)$$

$$k_4 = hf(\mathbf{x}_0 + k_3, t_0 + h)$$

$$\mathbf{x}(t_0 + h) = \mathbf{x}_0 + \frac{1}{6}k_1 + \frac{1}{3}k_2 + \frac{1}{3}k_3 + \frac{1}{6}k_4$$

Adaptive Stepsizes



- If we choose a fixed stepsize, we can only proceed as fast as the worst sections of $\mathbf{x}(t)$ will allow. By measuring an error e , we can change stepsize adaptively in runtime

$$e = |\mathbf{x}_a - \mathbf{x}_b|$$

\mathbf{x}_a : an estimate by taking an Euler step of size h

\mathbf{x}_b : an estimate by taking an Euler step of size $h/2$

- Suppose that we are willing to have an error of as much as 10^{-4} per step, where $\mathbf{x}_a, \mathbf{x}_b$ should differ from each other by $O(h^2)$

- If the current error is 10^{-8} $\left(\frac{10^{-4}}{10^{-8}}\right)^{1/2} h = 100h$  Can increase timestep upto 100
- If the current error is 10^{-3} $\left(\frac{10^{-4}}{10^{-3}}\right)^{1/2} h \approx .316h$  Should decrease timestep by 0.316 at least

Summary

- We have learned the basics of ordinary differential equations and some numerical methods for solving the equations
- In the upcoming lectures, we will use those methods to simulate many natural phenomena by integrating ODEs
- Euler's method is rarely used in real applications but it provides a foundation of numerical approaches